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ABSTRACT

Reporting and interpretation of effect sizes and structure coefficients in multiple regression results are important for good practice. The purpose of this study was to investigate the use and interpretation of effect sizes (ES) and structure coefficients in multiple regression analyses in two mathematics and science education journals. Published studies in five complete volumes of the "Journal for Research in Mathematics Education" and "School Science and Mathematics" were inspected. Nineteen multiple regression research articles were located and coded. Results suggest that effects are generally reported by researchers using multiple regression, but structure coefficients are seldom examined to help evaluate variable importance. (Contains 41 references.) (Author/SLD)

Running head: EFFECT SIZES AND STRUCTURE COEFFICIENTS

Use and Interpretation of Effect Sizes and Structure Coefficients
in Mathematics and Science Education Journals

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Abstract

Reporting and interpretation of effect sizes and structure coefficients in multiple regression results is important for good practice. The purpose of this paper is to investigate the use and interpretation of effect size (ES) and structure coefficients in multiple regression analyses in two mathematics and science education journals. Published studies in the Journal for Research in Mathematics Education and School Science and Mathematics were inspected. Results suggest that effects are generally reported by researchers using multiple regression, but structure coefficients are seldom examined to help evaluate variable importance.

Use and Interpretation of Effect Sizes and Structure Coefficients in Mathematics and Science Education Journals

Multiple regression analysis is a common tool used in social science research. Multiple regression can be useful in both predictive and explanatory research (Pedhazur, 1997). More than three decades ago, Jacob Cohen (1968) demonstrated that multiple regression was the univariate general linear model (GLM) and subsumed other univariate methods as special cases. Knapp (1978) later demonstrated canonical correlation as the multivariate GLM (see also Bagozzi, Fornell & Larcker, 1981; Henson, 2000; Thompson 1991).

Because all GLM analyses are correlational in nature, they all yield r^2 -type effect sizes that should be both reported and interpreted (cf. Cohen, 1994; Henson & Smith, 2000; Thompson, 1996). They also invoke weights (e.g., beta weights) that are applied to observed variables to create synthetic variables (e.g. \hat{Y} predicted scores) that become the focus of the analysis. Therefore, the relationship between the observed variables and the synthetic variables becomes important in determining the value of the observed variable. The correlation between an observed and synthetic variable is called a structure coefficient, which are also present in all GLM analyses (Courville & Thompson, 2001; Thompson & Borrello, 1985). In regression, for example, the correlation (structure coefficient) between a predictor and the \hat{Y} predicted scores informs the researcher as to the potential contributory value of the predictor to the effect observed.

Further, within any GLM analysis such as multiple regression, researchers are typically faced with a two-stage process of result interpretation. Stage one concerns whether the researcher has a result that is noteworthy, which historically has been evaluated with statistical significance testing. More recently, the role and import of effect sizes have seen increased use,

including a mandate to “always” report effects because such reporting is “essential” to good research practice (Wilkinson & APA Task Force on Statistical Inference, 1999, p. 599).

Once a noteworthy effect is observed, then (and only then), the researcher likely is concerned with determining what variables contributed to the effect. This is typically done by looking at standardized weights, but should also include examination of structure coefficients. For example, in multiple regression, a noteworthy R^2 effect is generally followed by study of the predictor beta weights (or unstandardized b weights) but should also invoke examination of the predictor structure coefficients. Courville and Thompson (2001) provide a thorough discussion on the definition and role of structure coefficients in multiple regression. Henson (2002) discusses structure coefficients in multivariate analyses.

Purpose

The purpose of the present paper is to investigate if suggested reporting standards of statistical information are being followed in two mathematics and science education journals. Focus was on the use and interpretation of effect size and structure coefficients in multiple regression applications. The literature suggests effect size reporting is generally sparse (DeVaney, 2001; Henson & Smith, 2000; Thompson, 2001), and the interpretation of effect sizes occurs even less often. However, Kirk (1996) observed higher rates of reporting for journals that frequently published articles employing multiple regression. Kirk postulated that effect reporting was greater for these journals because statistical software packages routinely provide R^2 effects in their outputs. This begs the question of whether researchers are thoughtfully including effects or if reporting is a function of what the computer output provides. Although an effect is reported, whether the effect is interpreted is another matter altogether.

Further, the reporting and interpretation of structure coefficients tends to be the exception rather than the rule. For the present study, I expected to observe moderate frequency of effect size reporting and very minimal use of structure coefficients.

Statistical Significance Testing and Effect Size

Statistical significant testing became common practice in the early to mid-1900's with its introduction into textbooks (DeVaney, 2001). Fan (2001) stated "..., statistical significance testing answers the question: What is the probability of obtaining an observed sample statistic ... when the population has a known parameter value ...?" (p. 275). Statistical significance testing is studied in most statistics courses, which often leaves graduate students with the idea that statistical significance testing is the most important part of the statistical interpretation. Further, the common p-value is routinely the only statistic interpreted in educational research (Plucker, 1997).

Statistical significance testing has supporters and critics. The supporters take the position that when used correctly, statistical significance testing describes valuable information (DeVaney, 2001; Henson & Smith, 2000). Critics express that statistical significance testing is often misinterpreted as practical significance (DeVaney, 2001; Henson & Smith, 2000). The dependence of statistical significance testing on sample size often confuses correct interpretations of statistical significance (DeVaney, 2001; Fan, 2001; Plucker, 1997). With the apparent limitations of statistical significance tests, effect sizes have been advocated for result interpretation (Fan, 2001).

Effect size in multiple regression typically refers to a variance-accounted-for statistic known as R^2 . Effect size does not replace statistical significance testing, and effect size reporting should complement statistical significance testing. After reporting the results of

significance testing, effect sizes should be reported and interpreted. Plucker (1997) reported that researchers in the social sciences are encouraged to provide effect size or the information for the calculation of the effect size in the results of studies. The fifth edition of the APA Publication Manual expects the reporting of effect size (APA, 2001). However, as Henson and Smith (2000) noted, “A more appropriate measure of effect size use would include an assessment of whether researchers both report and interpret their obtained effect sizes” (p. 290, emphasis added).

Henson and Smith (2000) suggested that the trend toward reporting of effect size is progressing slowly. Even with the encouragement to report effect size by fourth edition of the APA Manual (APA, 1994), the reporting of effect size is often neglected (DeVaney, 2001). Eleven empirical studies have been completed that show this encouragement by the APA has been ineffective (Thompson, 2001).

Therefore, one purpose of this research seeks to determine the extent to which authors of mathematics and science education studies report and interpret effect size.

Beta Weights and Structure Coefficients

After finding noteworthy effects, the researcher is charged with determining the variables that contributed to those effects. In regression and other GLM analyses, this determination should invoke study of both standardized weights and structure coefficients.

Beta weights are standardized coefficients (i.e., regression coefficients generated from standardized data) that suggest the independent variable’s predictive contribution to the dependent variable. Beta weights are easy to determine as they are generated in the output of multiple regression computer programs. However, beta weights are affected by collinearity, when predictor variables are correlated with each other. Hence, beta weights may yield distorted interpretations of variable import (Thompson & Borrello, 1985).

Structure coefficients in multiple regression are the correlations between the predictor variable and the estimated (predicted) score in the regression analysis (i.e., \hat{Y}). Structure coefficients (in squared format) express how much variance in the predicted variable (\hat{Y}) that the predictor variable possibly could explain. Because they are simple bivariate correlations, structure coefficients are not affected by collinearity as betas are (Thompson & Borrello, 1985).

Because beta weights may not express accurate variable interpretation, structure coefficients are being supported as an important step in statistical interpretation (Courville & Thompson, 2000; Thompson, 2001). Cooley and Lohnes (Thompson, 1990) argue for the reporting of structure coefficients by the general researcher. Of course, structure coefficients yield the same interpretation as zero-order correlations between the predictor and the dependent variable (Knapp, 1978). However, structures are interpretation in terms of the synthetic \hat{Y} variable.

Method

Multiple regression studies obtained from the Journal for Research in Mathematics Education and the School Science and Mathematics journal were identified. Every journal article of each issue was inspected for articles using multiple regression. I began with the latest issue and continued back until five complete volumes of each journal were exhausted. For the Journal of Research in Mathematics Education, I began with Volume 32, Number 4 and stopped after the inspection of Volume 27. For School Science and Mathematics, I began with Volume 101, Number 7 and stopped after completing volume 96. After a multiple regression analysis was located, the article was read for the reporting and interpretation of effect sizes, Pearson correlation coefficients (r), and structure coefficients. This information was coded in a table. Also, the type of effect size was noted.

The first multiple regression from each article was selected for inspection. A study received credit for reporting the various information if the information was described in tables or in words. A study received credit for interpretation of effect sizes, Pearson correlation coefficient, and structure coefficients if an explicit description of or reference to the statistic was included.

Results

Nineteen multiple regression research articles were located and coded. Simple frequencies for observations are presented in Table 1.

Table 1.

Summary of Reporting and Interpretation in Multiple Regression Articles.

	Report			Interpret		
	Effect Size	Pearson r	Structure Coefficients	Effect Size	Pearson r	Structure Coefficients
Yes	16 (84%)	8 (42%)	1 (5%)	16 (84%)	9 (47%)	1 (5%)
No	3 (16%)	11 (58%)	18 (95%)	3 (16%)	10 (53%)	18 (95%)
Total	19	19	19	19	19	19

Approximately 84% of the articles reviewed reported effect size. The interpretation of the effect size was the same at approximately 84% of the studies. Every article that reported effect size, interpreted the effect size. Three studies did not report or interpret effect size. When the effect size was reported, R^2 was the statistic of choice in all cases.

Pearson correlation coefficients (r) were reported in approximately 42% of the multiple regression articles. All articles that reported the Pearson correlation coefficient, also interpreted

the coefficient. One study interpreted Pearson correlation coefficients but did not report the statistic (Schwartz, 2000).

The use of structure coefficients was as expected and almost non-existent. Only one multiple regression article reported and interpreted structure coefficients (Terry & Baird, 1997).

Discussion

These results of the investigation of multiple regression analyses in mathematics and science education journals disagree with other studies that suggest the reporting of effect size to be slowly occurring (Henson & Smith, 2000) and often neglected (DeVaney, 2001). When using multiple regression analyses, mathematics and science researchers appear to have been responsive to the recommendations of the APA Publication Manual (APA, 2001). However, as Kirk (1996) suggested effect reporting is likely higher for regression research due to the commonly known \underline{R}^2 effect size routinely provided by statistical software outputs.

In this review of the literature, the only effect size reported was \underline{R}^2 . This is consistent with a report by Fan (2001) that \underline{R}^2 is the most popular effect size reported in several journals of psychology and Kirk's (1996) review. Statistical software packages routinely present this information in their regression output, making \underline{R}^2 readily available. Therefore, when regression is used, effect reporting tends to be more frequent. However, for analyses that statistical packages tend not to output effects, reporting is more sparse. This reality begs the question of whether effect reporting is a function of thoughtful researcher judgment or statistical output.

Adjusted \underline{R}^2 is also printed in the regression output of most statistical packages. However, adjusted \underline{R}^2 was not reported in the current study. Adjusted \underline{R}^2 is a "corrected" effect and will be less than or equal to the \underline{R}^2 . Corrected effects correct for sampling error and are better estimates of the true population effect.

Unfortunately, mathematics and science education researchers seldom examine structure coefficient to determine predictor importance. Zero-order correlations are also seldom interpreted beyond beta weights or unstandardized coefficients. Because statistical packages do not routinely report these coefficients in their outputs, structure coefficients are not as easily found (although they may be found by simply dividing the bivariate correlation between a predictor and the dependent variable by the multiple R). Because of this lack of reporting and interpretation, it is certainly possible that the researchers made erroneous conclusions about what predictors could have predicted the dependent variable. Courville and Thompson (2001) provide several examples of such misstatements.

In sum, good research multiple regression practice should consider effect sizes when deciding if a result is noteworthy and both standardized weights and structure coefficients when deciding what predictors were valuable in generating the observed effect. The present review of two mathematics and science education journals suggest reasonable reporting of effects, with an equal amount of interpretation of those effects, and almost non-existent examination of structure coefficients.

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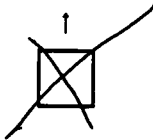
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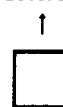
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